

WEEK - 13

Aim: to analyze and understand the architecture and working principle of a pre-trained deep learning model (such as VGG16, ResNet, Inception) used for image classification and feature extraction.

Algorithm:

1. Import a pre-trained model from keras applications.
2. Load model with pretrained weights
3. Display model summary (layers, parameters)
4. Analyze layer types (conv, pool, dense)
5. Identify trainable vs non-trainable parameters
6. optionally, visualize feature map.

pseudo code:

Begin

Import deep learning library (eg: tensorflow, PyTorch)

load a pre trained model:

→ include weights = 'imagenet'

→ exclude top layer if using for feature extraction.

Display model summary

→ print layer names, types, output shapes & parameters counts

for each layer in model:

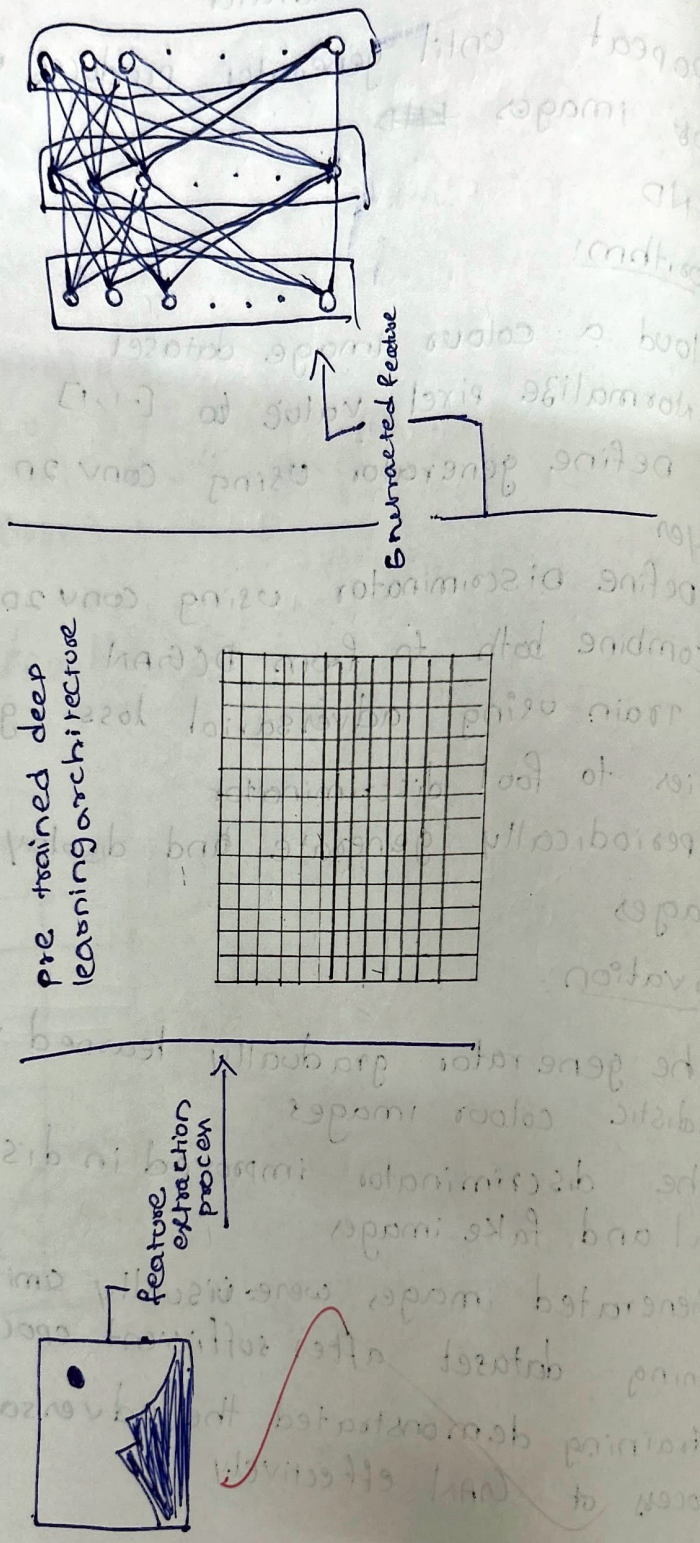
→ Identify type (conv, pool, dense, etc...)

→ Note activation function and no. of filters
visualize architecture diagram

→ Show flow from input image to output class
train the model

freeze lower layer if needed.

usage of pre-trained Architecture.

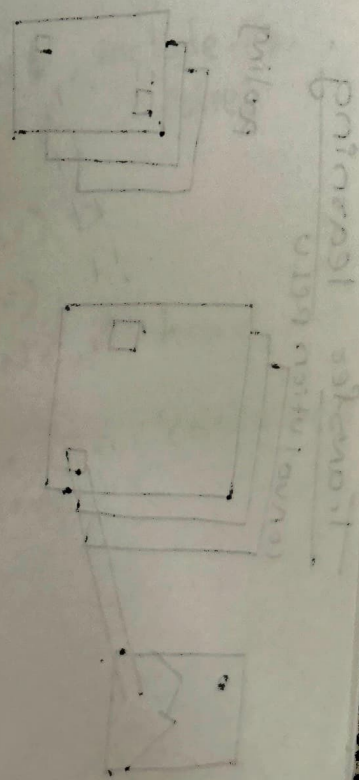
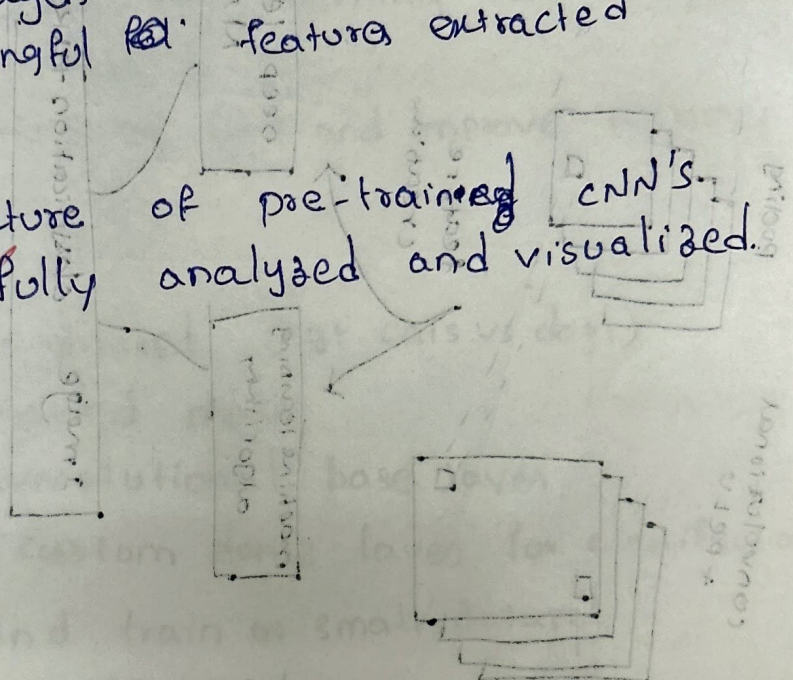


Observation:

- the pre-trained model has convolutional, pooling and fully connected layers
- convolutional layer extract features, pooling layers reduce size.
- pre-trained weights allow faster training and better accuracy
- sample images are correctly classified (or) have meaningful features extracted

Result:

the Architecture of pre-trained CNN's was successfully analyzed and visualized.



output

Total parameters : 138357544

Trainable parameters: 138357544

Non Trainable parameters : 0

layers Names:

convol : conv2d

bn1 : Batch Normal 2D

maxpool : Relu

layer1 : sequential

layer2 : sequential

layer3 : sequential

layer4 : sequential

avg pool : adaptive avg pool2d

fc : linear.

▶ # Lab 13: Understanding the architecture of a pre-trained model (VGG16)

```
import tensorflow as tf
from tensorflow.keras.applications import VGG16

# Load pre-trained VGG16 model with ImageNet weights
model = VGG16(weights="imagenet", include_top=True)

# Display model summary
model.summary()

# Plot model architecture
tf.keras.utils.plot_model(model, to_file="vgg16_architecture.png", show_shapes=True, show_layer_names=True)

print("\n✅ Model architecture plot saved as 'vgg16_architecture.png'")
```

📄 Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/vgg16/vgg16_weights_tf_dim_ordering_tf_kernels.h5
553467096/553467096 — 3s 0us/step
Model: "vgg16"

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/vgg16/vgg16_weights_tf_dim_ordering_tf_kernels.h5
553467096/553467096 3s 0us/step
Model: "vgg16"

Layer (type)	Output Shape	Param #
input_layer_13 (InputLayer)	(None, 224, 224, 3)	0
block1_conv1 (Conv2D)	(None, 224, 224, 64)	1,792
block1_conv2 (Conv2D)	(None, 224, 224, 64)	36,928
block1_pool (MaxPooling2D)	(None, 112, 112, 64)	0
block2_conv1 (Conv2D)	(None, 112, 112, 128)	73,856
block2_conv2 (Conv2D)	(None, 112, 112, 128)	147,584
block2_pool (MaxPooling2D)	(None, 56, 56, 128)	0
block3_conv1 (Conv2D)	(None, 56, 56, 256)	295,168
block3_conv2 (Conv2D)	(None, 56, 56, 256)	590,080
block3_conv3 (Conv2D)	(None, 56, 56, 256)	590,080
block3_pool (MaxPooling2D)	(None, 28, 28, 256)	0
block4_conv1 (Conv2D)	(None, 28, 28, 512)	1,180,160
block4_conv2 (Conv2D)	(None, 28, 28, 512)	2,359,808
block4_conv3 (Conv2D)	(None, 28, 28, 512)	2,359,808
block4_pool (MaxPooling2D)	(None, 14, 14, 512)	0
block5_conv1 (Conv2D)	(None, 14, 14, 512)	2,359,808
block5_conv2 (Conv2D)	(None, 14, 14, 512)	2,359,808
block5_conv3 (Conv2D)	(None, 14, 14, 512)	2,359,808
block5_pool (MaxPooling2D)	(None, 7, 7, 512)	0
flatten (Flatten)	(None, 25088)	0
fc1 (Dense)	(None, 4096)	102,764,544
fc2 (Dense)	(None, 4096)	16,781,312
predictions (Dense)	(None, 1000)	4,097,000

Total params: 138,357,544 (527.79 MB)

Trainable params: 138,357,544 (527.79 MB)

Non-trainable params: 0 (0.00 B)

Model architecture plot saved as 'vgg16_architecture.png'